

Midterm Feedback



- Mini-HW
- NTU COOL
- TA hours
- Course recordings
- Instant feedback



- Classroom (crowded, sleepy, etc.)
- Homework due time
- Pseudo code
- Difficulty of homework & exam
- TA recitation
- Seat announcement



Announcement

- Mini-HW 7 released
 - Due on 11/29 (Thur) 14:20
- Homework 3 released soon
 - Due on 12/13 (Thur) 14:20 (three weeks)

Frequently check the website for the updated information!



Mini-HW 7

Given a tree with N nodes, where each edge of the tree is weighted with W_i .

(1) Please design an algorithm (that runs in O(N) time) to accumulate the weights of all edges linking u and v. (For this question, a clear explanation is enough, no pseudo code is needed)

(2) Please simply justify the correctness of your algorithm.

Outline



- Graph Basics
- Graph Theory
- Graph Representations
- Graph Traversal
 - Breadth-First Search (BFS)
 - Depth-First Search (DFS)
- DFS Applications
 - Connected Components
 - Strongly Connected Components
 - Topological Sorting



- A graph G is defined as G = (V, E)
 - V: a finite, nonempty set of vertices
 - E: a set of edges / pairs of vertices



 $V = \{1, 2, 3, 4, 5\}$ $E = \{(1, 2), (1, 3), (1, 4), (2, 4), (2, 5), (4, 5)\}$



- Graph type
 - Undirected: edge (u, v) = (v, u)
 - **Directed**: edge (u, v) goes from vertex u to vertex v; $(u, v) \neq (v, u)$
 - Weighted: edges associate with weights







How many edges at most can a undirected (or directed) graph have?

- Adjacent (相鄰)
 - If there is an edge (u, v), then u and v are adjacent.
- Incident (作用)
 - If there is an edge (u, v), the edge(u, v) is incident from u and is incident to v.
- Subgraph (子圖)
 - If a graph G' = (V', E') is a subgraph of G = (V, E), then $V' \subseteq V$ and $E' \subseteq E$



Degree

- The degree of a vertex *u* is the number of edges incident on *u*
 - In-degree of u: #edges (x, u) in a directed graph
 - Out-degree of u: #edges (u, x) in a directed graph
 - Degree = in-degree + out-degree
 - Isolated vertex: degree = 0

$$|E| = \frac{(\sum_i d_i)}{2}$$



Path

- a sequence of edges that connect a sequence of vertices
- If there is a path from u (source) to v (target), there is a sequence of edges (u, i₁), (i₁, i₂), ..., (i_{k-1}, i_k), (i_k, v)
- **Reachable**: v is reachable from u if there exists a path from u to v

Simple Path

All vertices except for u and v are all distinct

Cycle

A simple path where u and v are the same

Subpath

A subsequence of the path



Connected

- Two vertices are connected if there is a path between them
- A connected graph has a path from every vertex to every other

Tree

a connected, acyclic, undirected graph

Forest

an acyclic, undirected but possibly disconnected graph





- <u>Theorem</u>. Let G be an undirected graph. The following statements are equivalent:
 - G is a tree
 - Any two vertices in G are connected by a unique simple path
 - G is connected, but if any edge is removed from E, the resulting graph is disconnected.
 - G is connected and |E| = |V| 1
 - G is acyclic, and |E| = |V| 1
 - G is acyclic, but if any edge is added to E, the resulting graph contains a cycle





Seven Bridges of Königsberg (七橋問題)

 How to traverse all bridges where each one can only be passed through once







Euler Path and Euler Tour (一筆畫問題)

- Euler path
 - Can you traverse each edge in a connected graph exactly once without lifting the pen from the paper?
- Euler tour
 - Can you finish where you started?





Euler Path and Euler Tour

Is it possible to determine whether a graph has an *Euler path* or an *Euler tour*, without necessarily having to find one explicitly?

- Solved by Leonhard Euler in 1736
- G has an Euler path iff G has exactly 0 or 2 odd vertices
- G has an Euler tour iff all vertices must be even vertices

Even vertices = vertices with even degrees Odd vertices = vertices with odd degrees





Hamiltonian Path

- Hamiltonian Path
 - A path that visits each vertex exactly once
- Hamiltonian Cycle
 - A Hamiltonian path where the start and destination are the same
- Both are NP-complete



Real-World Applications

Modeling applications using graph theory

- What do the vertices represent?
- What do the edges represent?
- Undirected or directed?







Graph Representations

Graph Representations

- How to represent a graph in computer programs?
- Two standard ways to represent a graph G = (V, E)
 - Adjacency matrix
 - Adjacency list



Adjacency Matrix

Adjacency matrix = V × V matrix A with A[u][v] = 1 if
 (u, v) is an edge



	1	2	3	4	5	6
1		1	1			
2	1			1	1	
3	1			1		1
4		1	1			1
5		1				
6			1	1		

- For undirected graphs, A is symmetric; i.e., $A = A^T$
- If weighted, store weights instead of bits in A



Complexity of Adjacency Matrix

- Space: $\Theta(n^2)$
- ${\scriptstyle \bullet}$ Time for querying an edge: $\Theta(1)$
- Time for inserting an edge: $\Theta(1)$
- Time for deleting an edge: $\Theta(1)$
- Time for listing all neighbors of a vertex: $\Theta(n)$
- Time for identifying all edges: $\Theta(n^2)$
- Time for finding in-degree and out-degree of a vertex?



Adjacency List

- Adjacency lists = vertex indexed array of lists
 - One list per vertex, where for u ∈ V, A[u] consists of all vertices adjacent to u





If weighted, store weights also in adjacency lists



Complexity of Adjacency List

- Space: $\Theta(m+n)$
- Time for querying an edge: $\Theta(\deg) \Rightarrow \Theta(\log \deg)$
- Time for inserting an edge: $\Theta(1) \quad \Rightarrow \Theta(\log \deg)$
- Time for deleting an edge: $\Theta(\deg) \Rightarrow \Theta(\log \deg)$
- Time for listing all neighbors of a vertex: $\Theta(\deg)$
- Time for identifying all edges: $\Theta(m+n)$
- Time for finding in-degree and out-degree of a vertex?



Representation Comparison

- Matrix representation is suitable for dense graphs
- List representation is suitable for sparse graphs
- Besides graph density, you may also choose a data structure based on the performance of other operations

	Space	Query an edge	Insert an edge	Delete an edge	List a vertex's neighbors	Identify all edges
Adjacency Matrix	$\Theta(n^2)$	$\Theta(1)$	$\Theta(1)$	$\Theta(1)$	$\Theta(n)$	$\Theta(n^2)$
Adjacanov List	$\Theta(m+n)$	$\Theta(\mathrm{deg})$	$\Theta(1)$	$\Theta(\deg)$	$\Theta(deg)$	$\Theta(m+n)$
Aujacency List	O(m + n)	$\Theta(\log \deg)$	$\Theta(\log \deg)$)	$\mathcal{O}(m+m)$	





Textbook Chapter 22 – Elementary Graph Algorithms

Graph Traversal

- From a source vertex, systematically follow the edges of a graph to visit all reachable vertices of the graph
- Useful to discover the structure of a graph
- Standard graph-searching algorithms
 - Breadth-First Search (BFS, 廣度優先搜尋)
 - Depth-First Search (DFS, 深度優先搜尋)





Breadth-First Search

Textbook Chapter 22.2 – Breadth-first search

Breadth-First Search (BFS)



Breadth-First Search (BFS)

- Input: directed/undirected graph G = (V, E) and source s
- Output: a breadth-first tree with root s (T_{BFS}) that contains all reachable vertices
 - v.d: distance from s to v, for all $v \in V$
 - Distance is the length of a shortest path in G
 - $v \cdot d = \infty$ if v is not reachable from s
 - v. d is also the depth of v in T_{BFS}
 - $v.\pi = u$ if (u, v) is the last edge on shortest path to v
 - u is v's predecessor in $T_{\rm BFS}$



Breadth-First Tree

```
Initially T<sub>BFS</sub> contains only s
As v is discovered from u, v and (u, v) are added to T<sub>BFS</sub>
T<sub>BFS</sub> is not explicitly stored; can be reconstructed from v.π
Implemented via a FIFO queue
Color the vertices to keep track of progress:

GRAY: discovered (first time encountered)
```

- BLACK: finished (all adjacent vertices discovered)
- WHITE: undiscovered

```
BFS(G, s)
  for each vertex u in G.V-{s}
                                   O(n
    u.color = WHITE
    u.d = \infty
    u.pi = NIL
  s.color = GRAY
  s.d = 0
  s.pi = NIL
  O = \{ \}
  ENQUEUE(Q, s)
  while O! = \{\}
    u = DEQUEUE(Q)
    for each v in G.Adj[u]
                              O(\deg(u))
      if v.color == WHITE
        v.color = GRAY
        v.d = u.d + 1
        v.pi = u
        ENQUEUE (Q, v)
    u.color = BLACK
```

 $\implies O\left(n + \sum_{u} (\deg(u) + 1)\right) = O(n + m)$

BFS Illustration



BFS Illustration







- Definition of δ(s, v): the shortest-path distance from s to v = the minimum number of edges in any path from s to v
 - If there is no path from s to v, then $\delta(s, v) = \infty$
- The BFS algorithm finds the shortest-path distance to each reachable vertex in a graph G from a given source vertex s ∈ V.



Lemma 22.1 Let G = (V, E) be a directed or undirected graph, and let $s \in V$ be an arbitrary vertex. Then, for any edge $(u, v) \in E$, $\delta(s, v) \leq \delta(s, u) + 1$.

Proof

*s-v*的最短路徑一定會小於等於*s-u*的最短路徑距離+1

- Case 1: u is reachable from s
 - s-u-v is a path from s to v with length $\delta(s, u) + 1$
 - Hence, $\delta(s, v) \le \delta(s, u) + 1$
- Case 2: u is unreachable from s
 - Then v must be unreachable too.
 - Hence, the inequality still holds.





Lemma 22.2 Let G = (V, E) be a directed or undirected graph, and suppose BFS is run on G from a given source vertex $s \in V$. Then upon termination, for each vertex $v \in V$, the value v.d computed by BFS satisfies $v.d \ge \delta(s, v)$.

Proof by induction

BFS算出的d值必定大於等於真正距離

Inductive hypothesis: $v.d \ge \delta(s, v)$ after n ENQUEUE ops

- Holds when n = 1: s is in the queue and $v \cdot d = \infty$ for all $v \in V\{s\}$
- After n + 1 ENQUEUE ops, consider a white vertex v that is discovered during the search from a vertex u

 $v.d = u.d + 1 \ge \delta(s, u) + 1$ (by induction hypothesis) $\ge \delta(s, v)$ (by Lemma 22.1)

Vertex v is never enqueued again, so v. d never changes again

Lemma 22.3

Suppose that during the execution of BFS on a graph G = (V, E), the queue Q contains the vertices $\langle v_1, v_2, ..., v_r \rangle$, where v_1 is the head of Q and v_r is the tail. Then, v_r , $d \leq v_1$, d + 1 and v_i , $d \leq v_{i+1}$, d for $1 \leq i < r$.

• Q中最後一個點的d值 ≤ Q中第一個點的d值+1

Proof by induction

• Q中第i個點的d值 $\leq Q$ 中第i+1點的d值

Inductive hypothesis: v_r . $d \le v_1$. d + 1 and v_i . $d \le v_{i+1}$. d after n queue ops

- Holds when $Q = \langle s \rangle$.
- Consider two operations for inductive step:
 - Dequeue op: when $Q = \langle v_1, v_2, ..., v_r \rangle$ and dequeue v_1
 - Enqueue op: when $Q = \langle v_1, v_2, ..., v_r \rangle$ and enqueue v_{r+1}

Inductive H1 $v_r.d \le v_1.d + 1$ (Q中最後一個點的d值 ≤ Q中第一個點的d值+1) hypothesis: H2 $v_i.d \le v_{i+1}.d, i = 1, \dots, r - 1$ (Q中第i個點的d值 ≤ Q中第i+1點的d值)

Dequeue op



 v_2

 v_{r-1}

 v_r

 v_{r+1}

 $v_r.d \leq v_1.d + 1$ (induction hypothesis H1) $\leq v_2.d + 1$ (induction hypothesis H2) \rightarrow H1 holds $v_i.d \leq v_{i+1}.d, i = 2, \cdots, r-1 \rightarrow$ H2 holds

> Let u be v_{r+1} 's predecessor, $v_{r+1}.d = u.d + 1$ Since u has been removed from Q, the new head v_1 satisfies $u.d \leq v_1.d$ (induction hypothesis H2) $v_{r+1}.d \leq u.d + 1 \leq v_1.d + 1 \rightarrow$ H1 holds $v_r.d \leq u.d + 1$ (induction hypothesis H1) $v_r.d \leq u.d + 1 = v_{r+1}.d$ $v_i.d = v_{i+1}.d, i = 1, \cdots, r \rightarrow$ H2 holds (38)

Corollary 22.4

Suppose that vertices v_i and v_j are enqueued during the execution of BFS, and that v_i is enqueued before v_j . Then $v_i d \le v_j d$ at the time that v_j is enqueued.

Proof

若
$$v_i$$
比 v_j 早加入queue → v_i . $d \le v_j$. d

- Lemma 22.3 proves that $v_i d \le v_{i+1} d$ for $1 \le i < r$
- Each vertex receives a finite *d* value at most once during the course of BFS
- Hence, this is proved.



Theorem 22.5 – BFS Correctness

Let G = (V, E) be a directed or undirected graph, and and suppose that BFS is run on G from a given source vertex $s \in V$.

- 1) BFS discovers every vertex $v \in V$ that is reachable from the source s
- 2) Upon termination, $v \cdot d = \delta(s, v)$ for all $v \in V$
- 3) For any vertex $v \neq s$ that is reachable from s, one of the shortest paths from s to v is a shortest path from s to v. π followed by the edge $(v.\pi, v)$

Proof of (1)

• All vertices v reachable from s must be discovered; otherwise they would have $v \cdot d = \infty > \delta(s, v)$. (contradicting with Lemma 22.2)



(2) $v.d = \delta(s, v) \ \forall \ v \in V$

- Proof of (2) by contradiction
 - Assume some vertices receive d values not equal to its shortest-path distance
 - Let v be the vertex with minimum δ(s, v) that receives such an incorrect d value; clearly v ≠ s
 - By Lemma 22.2, $v.d \ge \delta(s, v)$, thus $v.d > \delta(s, v)$ (v must be reachable)
 - Let u be the vertex immediately preceding v on a shortest path from s to v, so $\delta(s, v) = \delta(s, u) + 1$
 - Because $\delta(s, u) < \delta(s, v)$ and v is the minimum $\delta(s, v)$, we have $u.d = \delta(s, u)$
 - $v.d > \delta(s,v) = \delta(s,u) + 1 = u.d + 1$



(2) $v.d = \delta(s, v) \ \forall \ v \in V$

- Proof of (2) by contradiction (cont.)
 - $v.d > \delta(s,v) = \delta(s,u) + 1 = u.d + 1$
 - When dequeuing u from Q, vertex v is either WHITE, GRAY, or BLACK
 - WHITE: v.d = u.d + 1, contradiction
 - BLACK: it was already removed from the queue
 - By Corollary 22.4, we have $v.d \le u.d$, contradiction
 - GRAY: it was painted GRAY upon dequeuing some vertex w
 - Thus v.d = w.d + 1 (by construction)
 - w was removed from Q earlier than u, so w. $d \le u. d$ (by Corollary 22.4)
 - $v.d = w.d + 1 \le u.d + 1$, contradiction
 - Thus, (2) is proved.



(3) For any vertex $v \neq s$ that is reachable from *s*, one of the shortest paths from *s* to *v* is a shortest path from *s* to *v*. π followed by the edge $(v.\pi, v)$

- Proof of (3)
 - If v. π = u, then v. d = u. d + 1. Thus, we can obtain a shortest path from s to v by taking a shortest path from s to v. π and then traversing the edge (v. π, v).



BFS Forest

- BFS (G, s) forms a BFS tree with all reachable v from s
- We can extend the algorithm to find a BFS forest that contains every vertex in G

```
//explore full graph and builds up
a collection of BFS trees
BFS(G)
for u in G.V
u.color = WHITE
u.d = ∞
u.π = NIL
for s in G.V
if(s.color == WHITE)
// build a BFS tree
BFS-Visit(G, s)
```

```
BFS-Visit(G, s)
s.color = GRAY
s.d = 0
s.π = NIL
Q = empty
ENQUEUE(Q, s)
while Q ≠ empty
u = DEQUEUE(Q)
for v in G.adj[u]
if v.color == WHITE
v.color = GRAY
v.d = u.d + 1
v.π = u
ENQUEUE(Q, v)
u.color = BLACK
```





Depth-First Search

Textbook Chapter 22.3 – Depth-first search

Depth-First Search (DFS)



DFS Algorithm

<pre>// Explore full graph and builds up</pre>
a collection of DFS trees
DFS (G)
for each vertex u in G.V $O(n)$
u.color = WHITE
u.pi = NIL
time = 0 // global timestamp
for each vertex u in G.V
if u.color == WHITE
DFS-VISIT(G, u)

```
DFS-Visit(G, u) O(deg(u) + 1)
time = time + 1
u.d = time // discover time
u.color = GRAY
for each v in G.Adj[u]
if v.color == WHITE
v.pi = u
DFS-VISIT(G, v)
u.color = BLACK
time = time + 1
u.f = time // finish time
```

• $O\left(n + \sum_{n \in I} (\deg(u) + 1)\right)$

- Implemented via recursion (stack)
- Color the vertices to keep track of progress:
 - GRAY: discovered (first time encountered)
 - BLACK: finished (all adjacent vertices discovered)
 - WHITE: undiscovered

47

= O(n+m)

Parenthesis Theorem

 Parenthesis structure: represent the discovery of vertex u with a left parenthesis "(u" and represent its finishing by a right parenthesis "u)". In DFS, the parentheses are properly nested.

White Path Theorem

- In a DFS forest of a directed or undirected graph G = (V, E),
 - vertex v is a descendant of vertex u in the forest ⇔ at the time u.d that the search discovers u, there is a path from u to v in G consisting entirely of WHITE vertices
- Classification of Edges in G
 - Tree Edge
 - Back Edge
 - Forward Edge
 - Cross Edge



Parenthesis Theorem

 Parenthesis structure: represent the discovery of vertex u with a left parenthesis "(u" and represent its finishing by a right parenthesis "u)". In DFS, the parentheses are properly nested.



- White Path Theorem
 - In a DFS forest of a directed or undirected graph G = (V, E),
 - vertex v is a descendant of vertex u in the forest ⇔ at the time u. d that the search discovers u, there is a path from u to v in G consisting entirely of WHITE vertices
- Proof.
 - >
 - Since v is a descendant of u, u. d < v. d
 - Hence, v is WHITE at time u. d
 - In fact, since v can be any descendant of u, any vertex on the path from u to v are WHITE at time u.d
 - ← (textbook p. 608)

- Classification of Edges in G
 - Tree Edge (GRAY to WHITE)
 - Edges in the DFS forest
 - Found when encountering a new vertex v by exploring (u, v)
 - Back Edge (GRAY to GRAY)
 - (u, v), from descendant u to ancestor v in a DFS tree
 - Forward Edge (GRAY to BLACK)
 - (*u*, *v*), from ancestor *u* to descendant *v*. Not a tree edge.
 - Cross Edge (GRAY to BLACK)
 - Any other edge between trees or subtrees. Can go between vertices in same DFS tree or in different DFS trees

In an undirected graph, back edge = forward edge.

To avoid ambiguity, classify edge as the first type in the list that applies.

- Edge classification by the color of v when visiting (u, v)
 - WHITE: tree edge
 - GRAY: back edge
 - BLACK: forward edge or cross edge
 - $u.d < v.d \rightarrow$ forward edge
 - $u.d > v.d \rightarrow cross edge$



Theorem 22.10

In DFS of an undirected graph, there are only tree edges and back edges without forward and cross edge.





DFS Applications

- Connected Components
- Strongly Connected Components
- Topological Sort





Connected Components

Connected Components Problem

- Input: a graph G = (V, E)
- Output: a connected component of G
 - a **maximal** subset U of V s.t. any two nodes in U are connected in G



Why must the connected components of a graph be disjoint?

Connected Components



Time Complexity: O(n+m)

BFS and DSF both find the connected components with the same complexity

Problem Complexity



Upper bound = O(m+n)

Lower bound = $\Omega(m+n)$







Question?

Important announcement will be sent to @ntu.edu.tw mailbox & post to the course website

Course Website: http://ada.miulab.tw

Email: ada-ta@csie.ntu.edu.tw